





## Prediction of Global Geomagnetic Fied Disturbances using Recurrent Neural Network

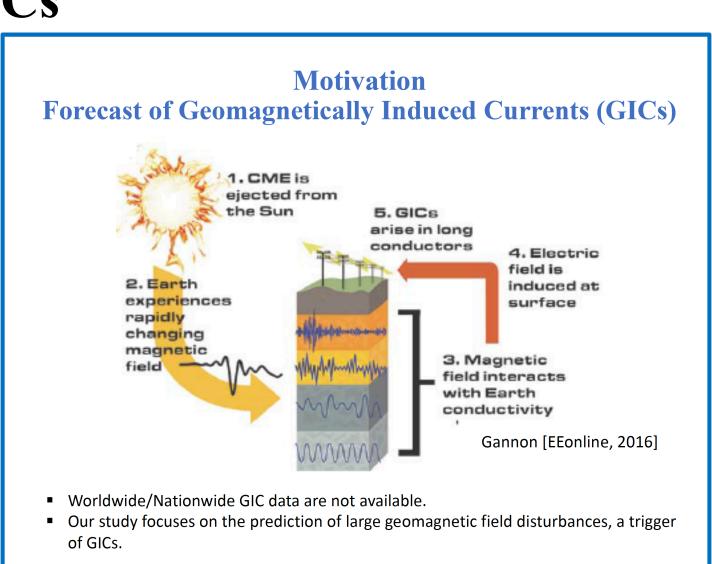
Hyunju Connor (<a href="https://linear.com/hkconnor@Alaska.edu">hkconnor@Alaska.edu</a>)<sup>1</sup>, Shishar Priyadarshi<sup>1</sup>, Matthew Blandin<sup>1</sup>, and Amy Keesee<sup>2</sup>

<sup>1</sup>University of Alaska Fairbanks, <sup>2</sup>University of New Hampshire

## **MAGICIAN** Team for Forecasting GICs

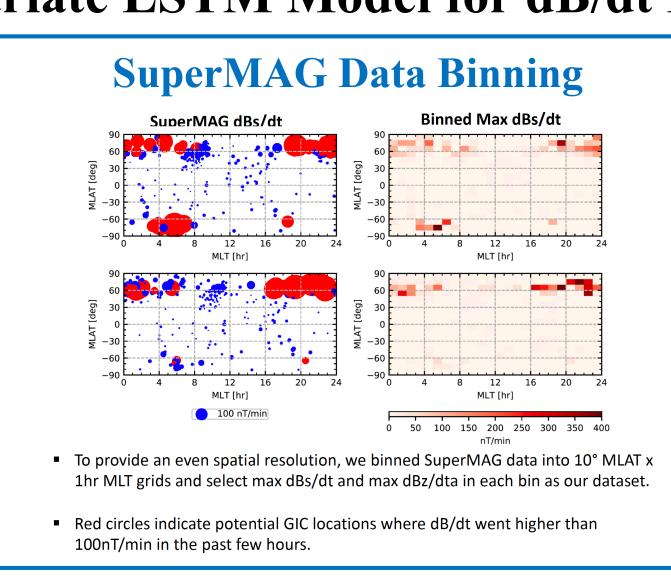
### **MAGICIAN Team** MAGICIAN is a joint UAF – UNH team funded by the 2018 NSF EPSCoR RII Track 2 Program that develops machine-learning algorithms for predicting hazardous Geomagnetically Induced Currents (GIC). **Machine Learning Algorithms for Geomagnetically Induced Currents** in Alaska and New Hampshire Check other presentations of MAGICIAN team at this AI & Data Science Workshop! Poster #53: Using an LSTM and Classification Methods to Determine Risk of dB/dt Threshold Crossings as Proxy for Geomagnetically Induced Currents -Michael Coughlan, UNH. Poster #57: Comparison of Time Series Techniques to Model Connections Between Solar Wind Input and Geomagnetically Induced Currents - Amy Keesee, UNH.



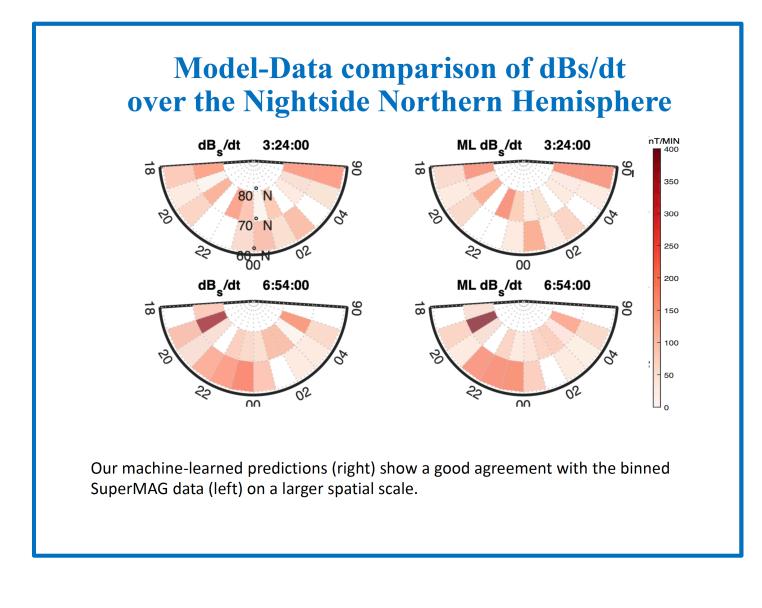


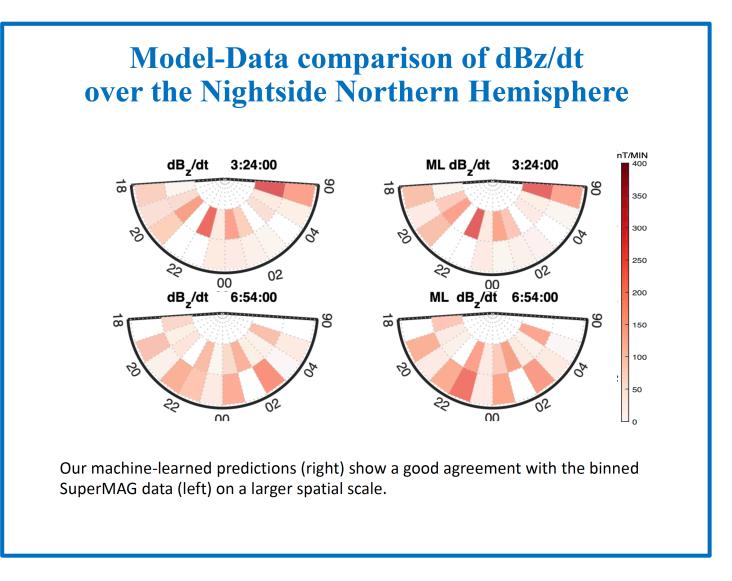
## Multi-variate LSTM Model for dB/dt Prediction

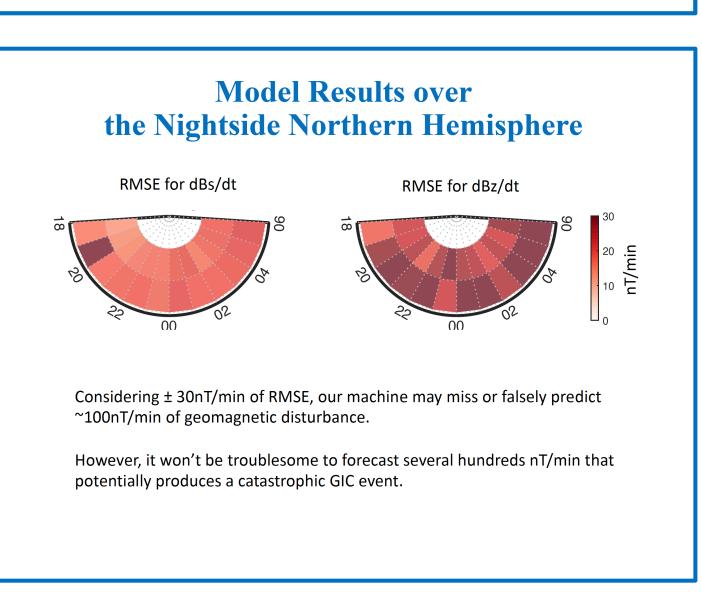
# Recurrent Neural Network (RNN) for predicting geomagnetic field disturbance (dB/dt) \*DATA: \*OMNI solar wind and IMF conditions in 2012 and 2015 \*SuperMAG surface & vertical disturbances (i.e., dBs/dt & dBz/dt) in 2012 and 2015 \*80% for training, 20% for validation, and the 2012-03-09 storm for testing \*Method: Multi-variate Long Short Term Memory (LSTM) network \*50 neurons in a single hidden layer, 50 epochs with a batch size of 72 \*Adam's stochastic gradient decent as an optimization algorithm \*Mean absolute error as a loss function \*Two machines are trained for dBs/dt and dBz/dt predictions. \*Input: IMF Bz, Solar Wind Density, dBs/dt (or dBz/dt) at a previous minute (t-1) \*Output: dBs/dt (or dBz/dt) at the next minute (t)



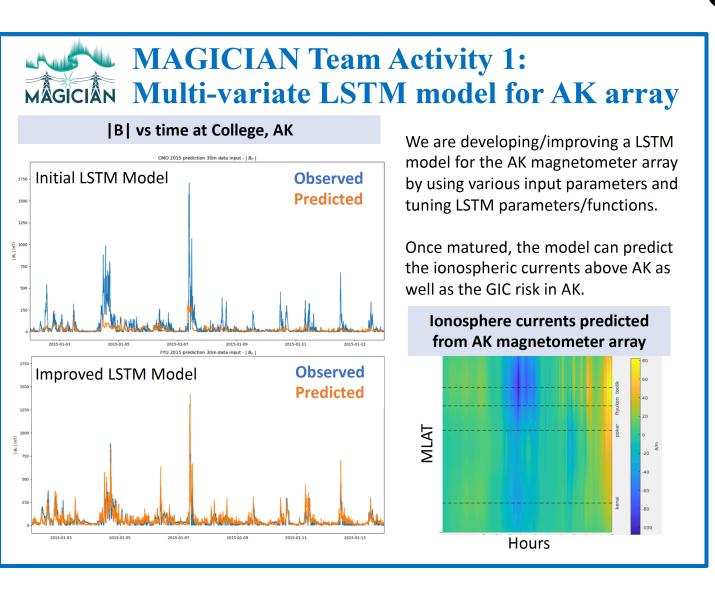
## Improvement needed in future Train machines with longer periods of data. Use a longer time history of input (e.g. 60 mins of SW/IMF data). Remove dBs/dt and dBz/dt from the input. Our current model may find stronger correlation with dB/dt at t-1min. Consider sophisticated solar wind propagation from the bow shock to each bins. 60min delay from the bow shock to the nightside bins were assumed. Consider finer spatial resolution for higher latitude Use better validation techniques than RMSE [Welling et al. SW2018; Maimaiti et al. SW2019; Camporeale. JGR 2020] Use different machine learning techniques Multi-layered LSTM, Artificial Neural Network, Convolutional Neural Network, Principal Component Analysis, etc.

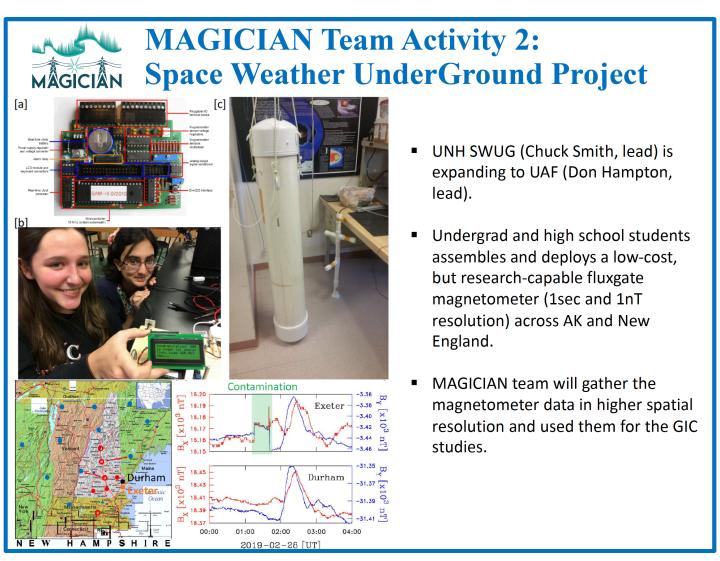






## Other MAGICIAN Team Activities







## Summary

- We developed a prototype of a multi-variate LSTM model using 2 years of OMNI and SuperMAG data.
- The prototype model catches over 100nT/min of dB/dt relatively well on 09 Mar 2012 geomagnetic storm.
- Once matured, this model can provide an advanced warning of GICs that are typically triggered by large dB/dt.
- In addition to the ML-GIC models, MAGICIAN team provides the low-cost, research-capable magnetometer arrays in AK and NH, and the GIC measurements in AK to the space science community.